

Animation in Computer Graphics

Presenter: Le Ngoc Hanh





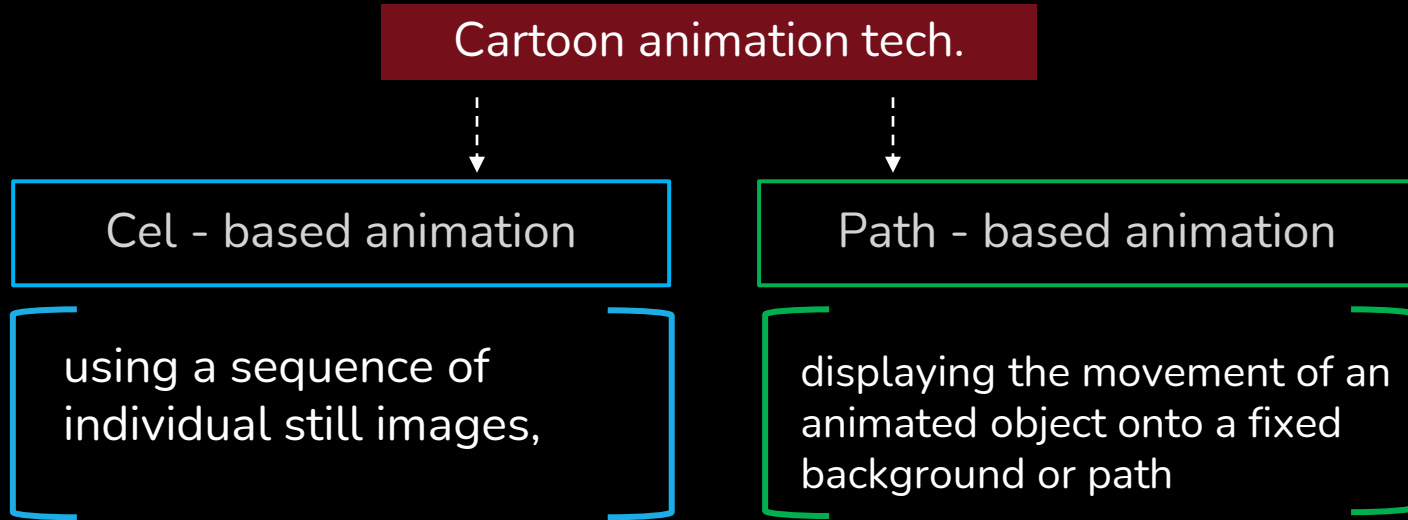
Outline

- ❖ Cartoon animation techniques
 - Golden age
 - Digital age
- ❖ Research side
- ❖ Text – driven animation

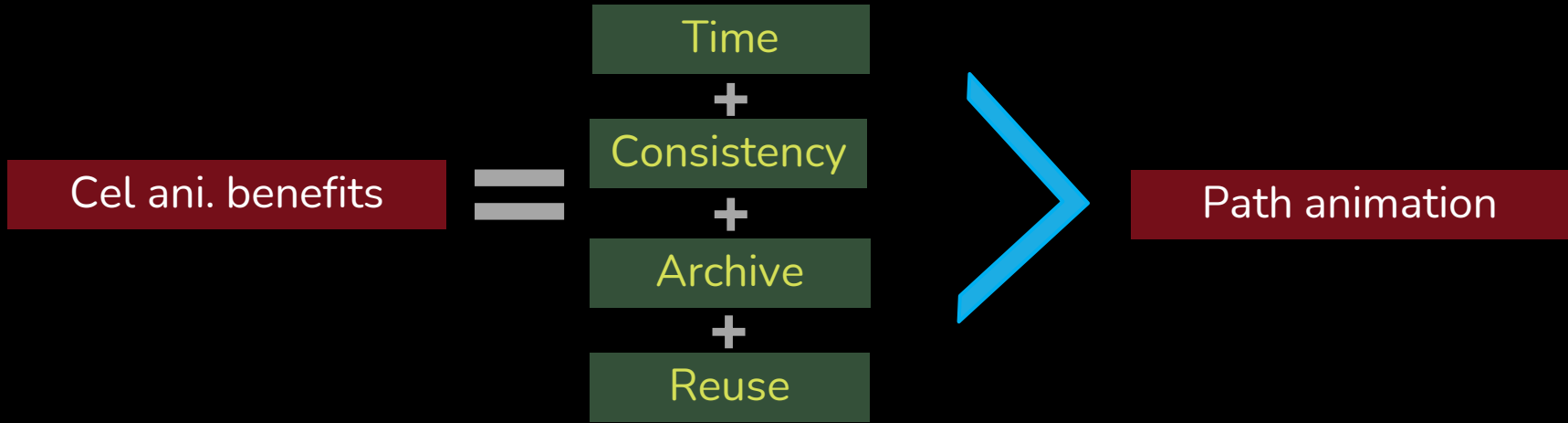
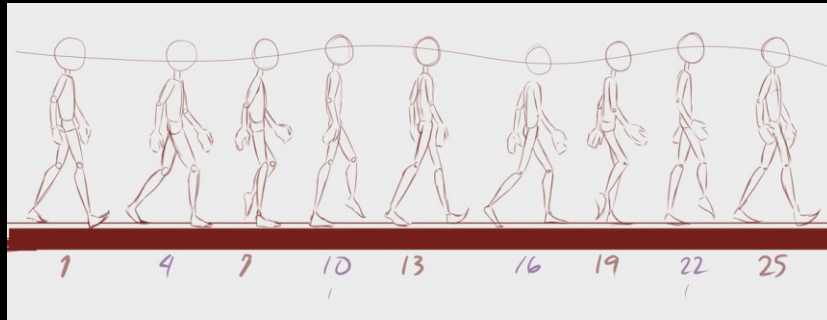




Cartoon animation techniques



Cartoon animation techniques



▲ Cel animation



▲ Cel animation



An animation cel from Snow White and the Seven Dwarfs • Disney cel animation



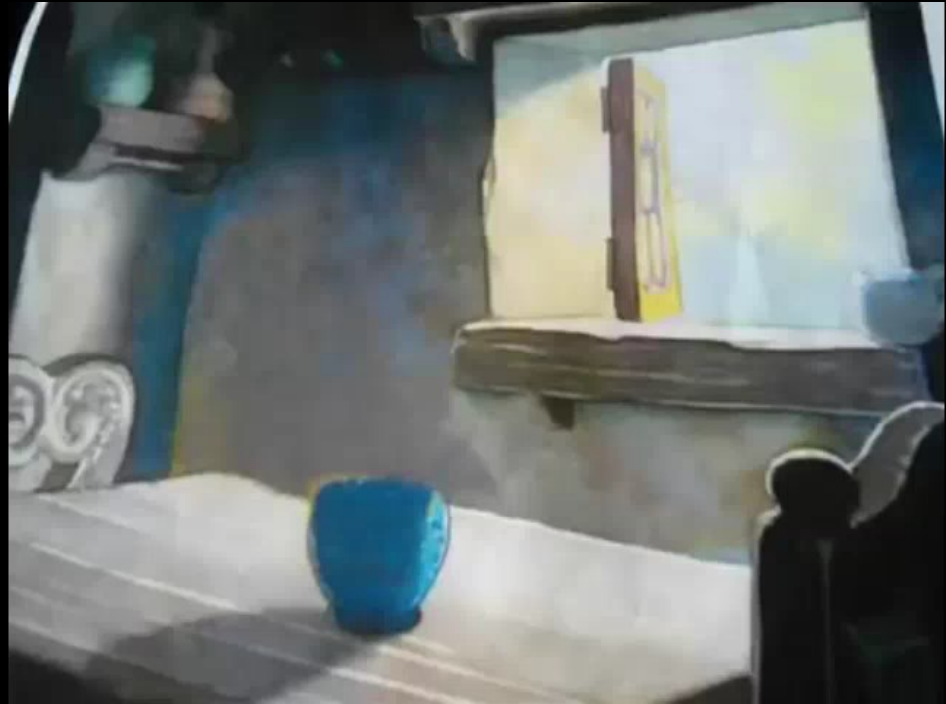
▲ Cel animation



▲ Cel animation

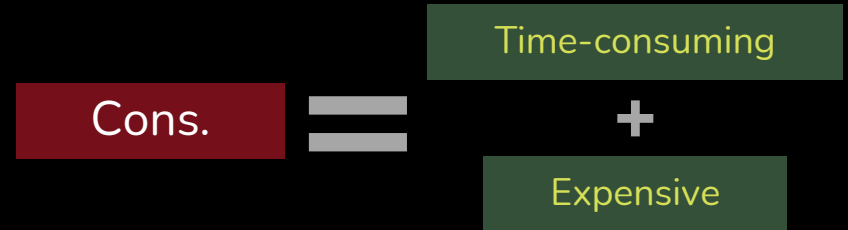
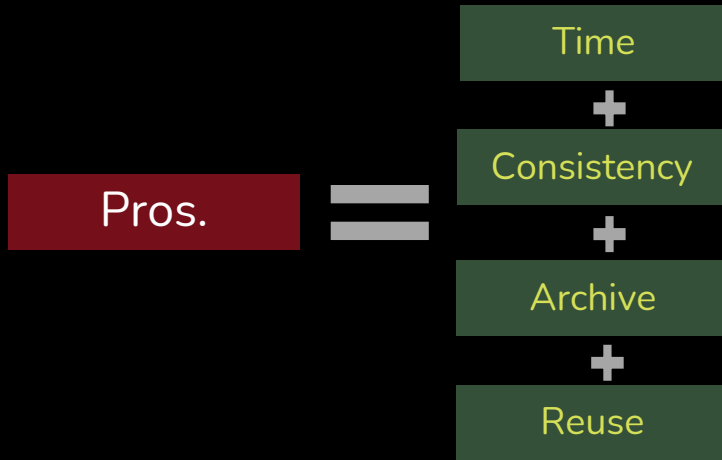
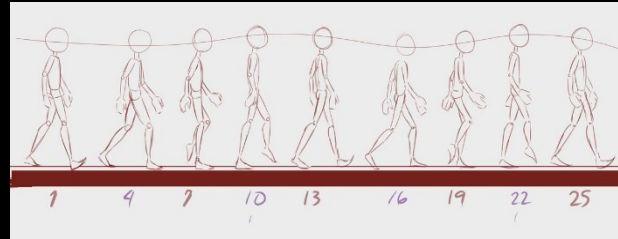


Cel / frame preparation



Create animation

Cartoon animation techniques

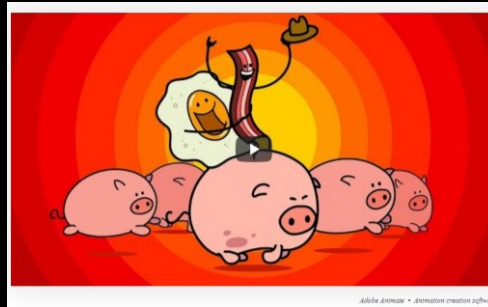


▲ Cel animation in the digital age

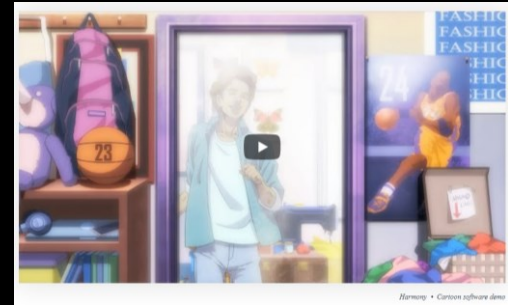
Adobe Animate



Adobe Character Animator



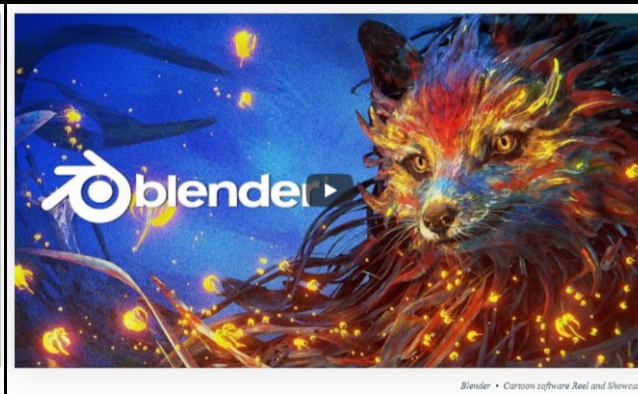
Toon Boom Harmony



Autodesk Maya



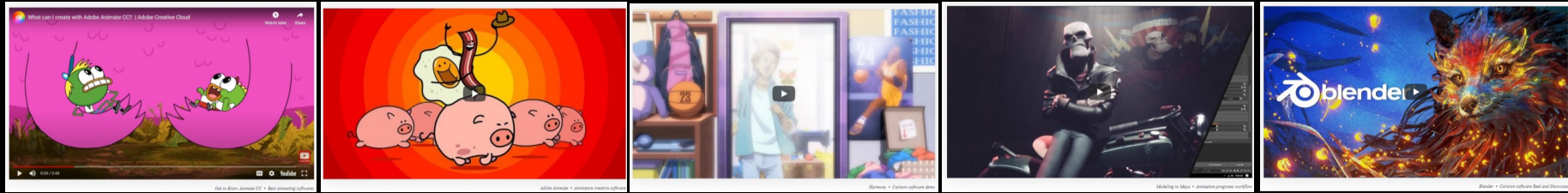
Blender



▲ Cel animation in the digital age



▲ Cel animation in the digital age



Basic animation



Regenerating Arbitrary Video Sequences with Distillation Path-Finding

Thi-Ngoc-Hanh Le, Shang-Yi Yao, Chun-Te Wu, and Tong-Yee Lee, *Senior Member, IEEE*

Cartoon Textures

Christina de Juan[†] and Bobby Bodenheimer[‡]

Vanderbilt University

On Combining Multiple Features for Cartoon Character Retrieval and Clip Synthesis

Jun Yu, Dongquan Liu, Dacheng Tao, *Senior Member, IEEE*, and Hock Soon Seah

Learning a perceptual manifold with deep features for animation video resequencing

Recognizing Cartoon Image Gestures for Retrieval and Interactive Cartoon Clip Synthesis

Yi Yang, Yueting Zhuang, *Member, IEEE*, Dacheng Tao, *Member, IEEE*, Dong Xu, *Member, IEEE*, Ji and Jiebo Luo, *Fellow, IEEE*

Charles C. Morace¹ · Thi-Ngoc-Hanh Le¹ · Sheng-Yi Yao¹ · Shang-Wei Zhang¹ · Tong-Yee Lee¹

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Eurographics/ACM SIGGRAPH Symposium on Computer Animation (2004)

R. Boulic, D. K. Pai (Editors)

Cartoon Textures

Christina de Juan[†] and Bobby Bodenheimer[‡]

Vanderbilt University

Research side of animation producing



Cartoon texture [a]

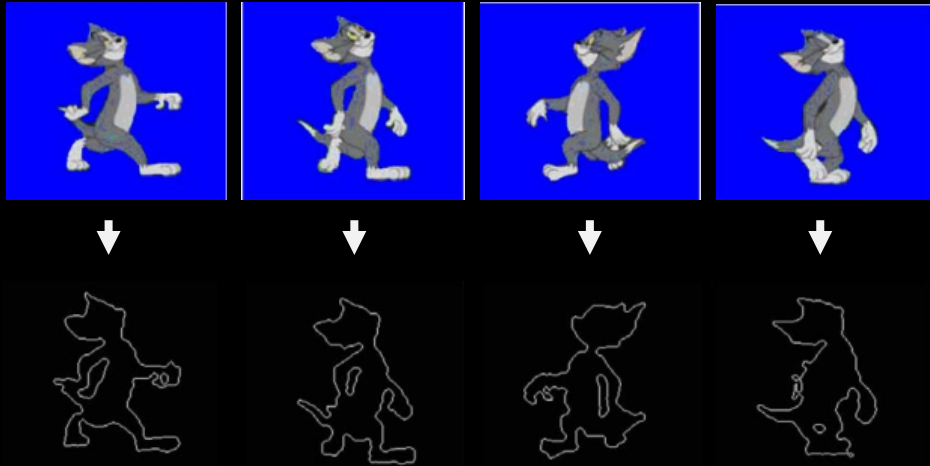
- ✖ Pre-processing is needed.
- ✖ Simple cartoon characters
- ✖ Simple motion

[a] de Juan, Christina, and Bobby Bodenheimer. "Cartoon textures." *Proceedings of the 2004 ACM SIGGRAPH/Eurographics symposium on Computer animation*. 2004.

Recognizing Cartoon Image Gestures for Retrieval and Interactive Cartoon Clip Synthesis

Yi Yang, Yueting Zhuang, *Member, IEEE*, Dacheng Tao, *Member, IEEE*, Dong Xu, *Member, IEEE*, Jun Yu,
and Jiebo Luo, *Fellow, IEEE*

Research side of animation producing



CGS [b]

- ✘ Need a pre-processing
- ✘ Motion analysis based on Gesture/ Shape of characters
- ✘ Simple motion
- ✘ Focus on cartoon characters

[b] Yang, Yi, et al. "Recognizing cartoon image gestures for retrieval and interactive cartoon clip synthesis." *IEEE transactions on circuits and systems for video technology* 20:12 (2010): 1745-1756.



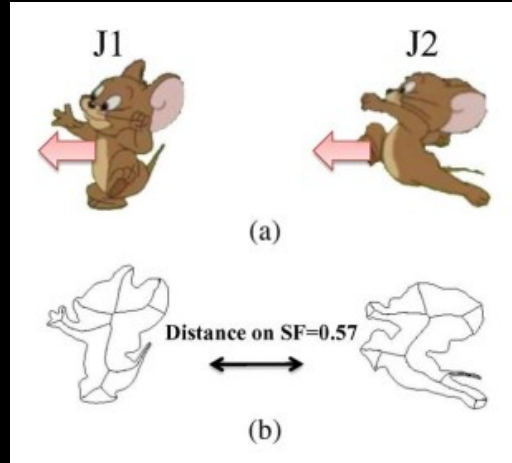
On Combining Multiple Features for Cartoon Character Retrieval and Clip Synthesis

Jun Yu, Dongquan Liu, Dacheng Tao, *Senior Member, IEEE*, and Hock Soon Seah

Research side of animation producing



Semi-MSL [c]



Simple motion

Some simple cartoon data

[c] Yu, Jun, et al. "On combining multiple features for cartoon character retrieval and clip synthesis." *IEEE Transactions on Systems, Man, and Cybernetics, Part B (Cybernetics)* 42.5 (2012): 1413-1427.

Learning a perceptual manifold with deep features for animation video resequencing

Charles C. Morace¹ · Thi-Ngoc-Hanh Le¹ · Sheng-Yi Yao¹ · Shang-Wei Zhang¹ ·
Tong-Yee Lee¹

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Research side of animation producing



Manifold sequence [d]

Heavily relying on the performance of an off-the-shelf distance metric LPIPS

Motion direction is not considered

Real-world videos with dense of motion → Fail

Resequence the source video, not from arbitrary frame.

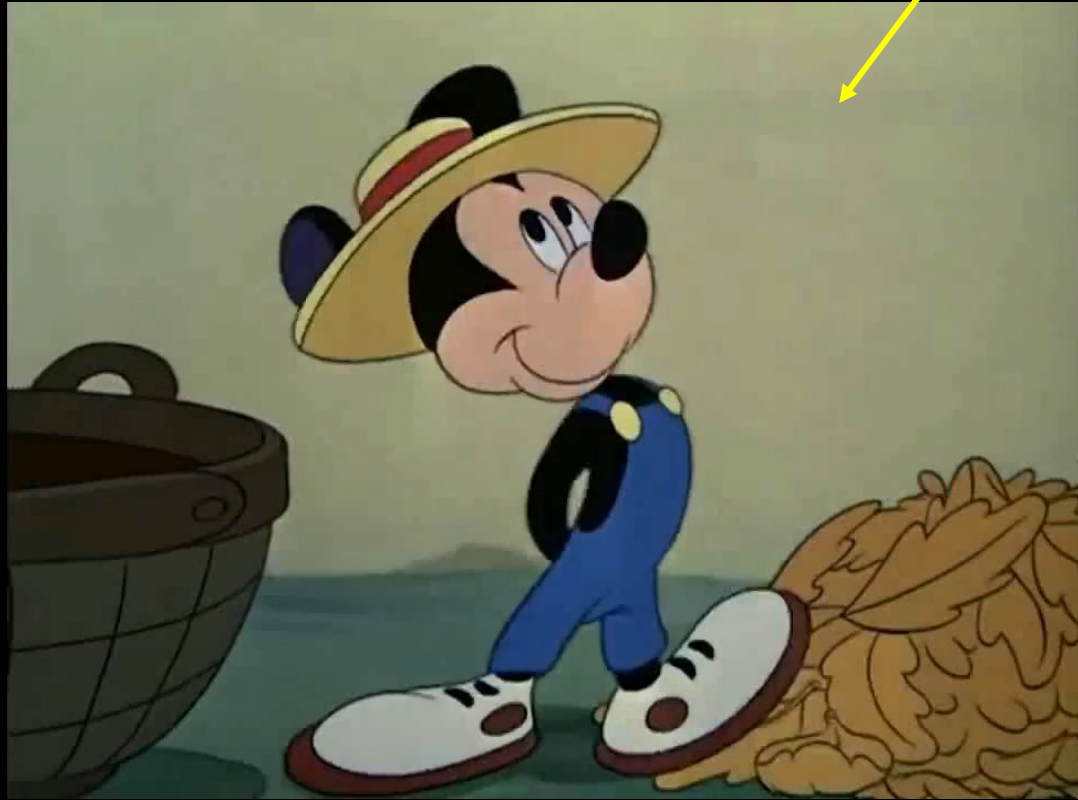
[d] Morace, Charles C., et al. "Learning a perceptual manifold with deep features for animation video resequencing." *Multimedia Tools and Applications* (2022): 1-21.

Research side of animation producing

Methods	Pre-processing	New sequence?	Type of data
GCCS [3]	Yes	No	Cartoon characters
RCCS [26]	Yes	No	Cartoon characters
semi-MSL [28]	Yes	No	Cartoon characters
Manifold [17]	No	No	Cartoon scenes

Deficiencies

- ❖ Linear motion segments
- ❖ Cartoon and natural scene
- ❖ Diversity in content and background.
- ❖ New sequences from arbitrary starting frame



Research side of animation producing

Methods	Pre-processing	New sequence?	Type of data
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Deficiencies

- ❖ Linear motion segments
- ❖ Cartoon and natural scene
- ❖ Diversity in content and background.
- ❖ New sequences from arbitrary starting frame

▷ Dense of moving objects in real-world videos



Input video

Output sequence

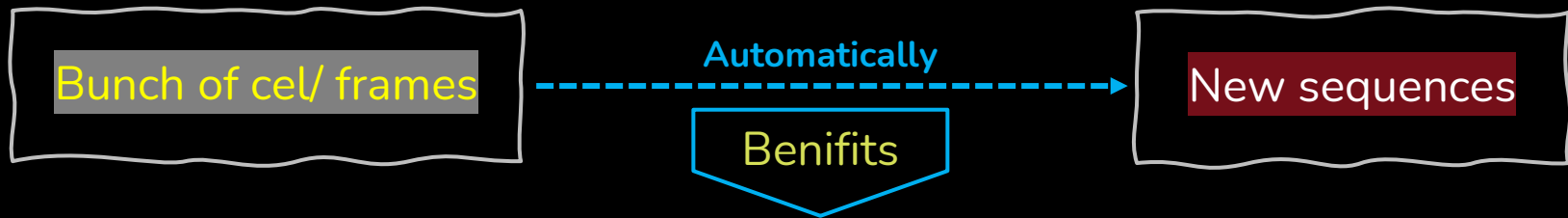
▶ Motions of both foreground and background



Input video

Output sequence

Research side of animation producing



⚛ Speed up the time of producing an art

⚛ Easily create variations of the existing material

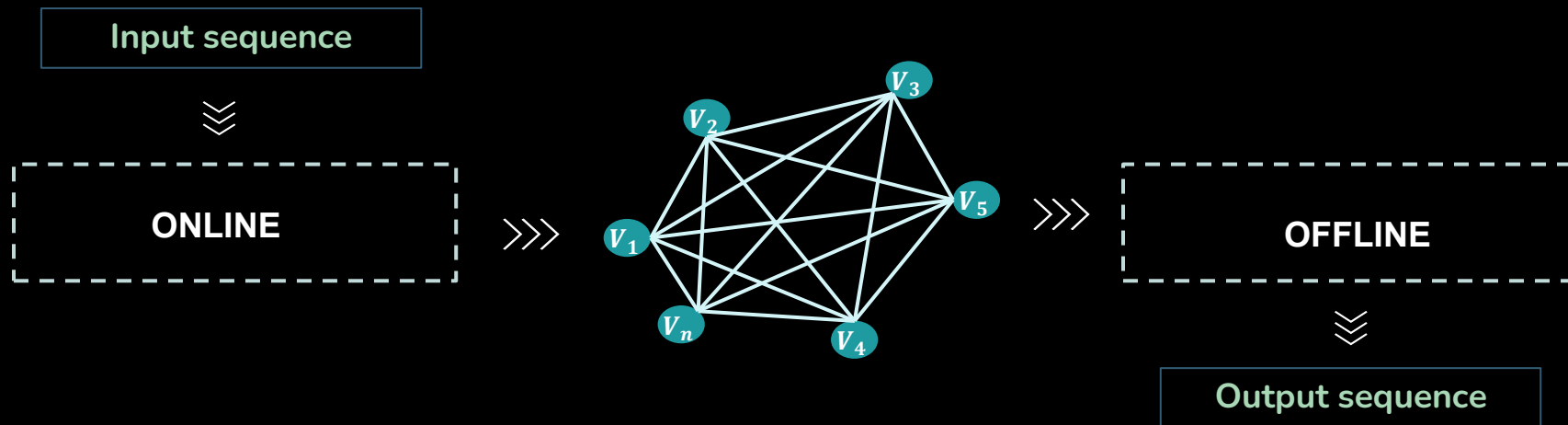
⚛ Expertise-free

IEEE TRANSACTIONS ON VISUALIZATION AND COMPUTER GRAPHICS

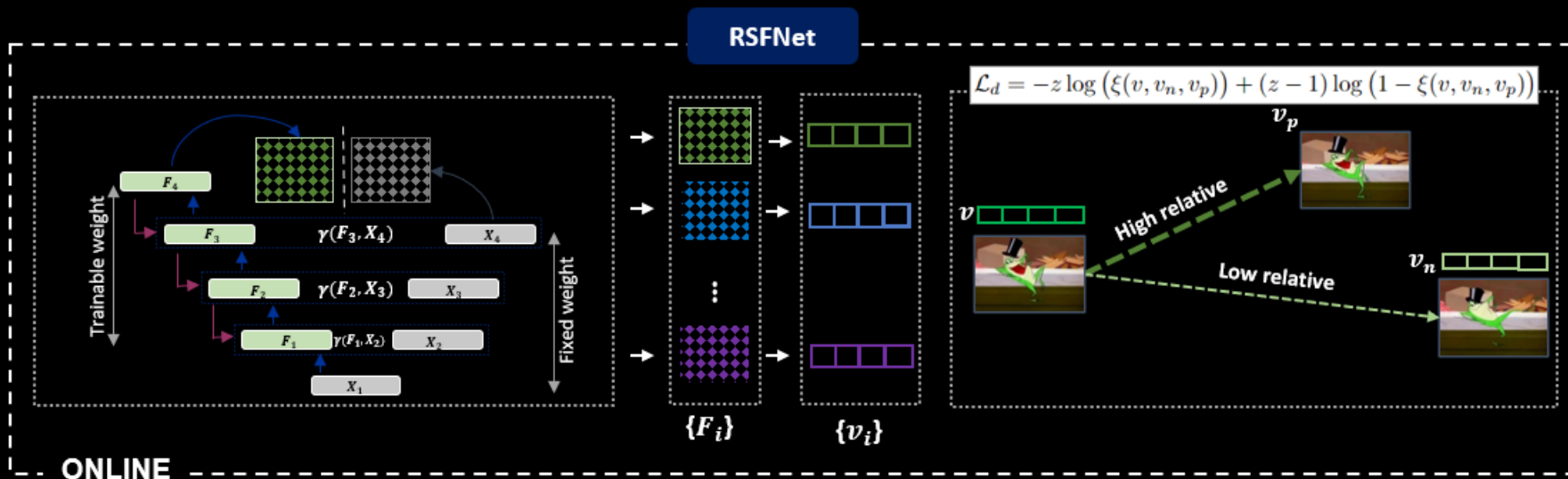
Regenerating Arbitrary Video Sequences with Distillation Path-Finding

Thi-Ngoc-Hanh Le, Shang-Yi Yao, Chun-Te Wu, and Tong-Yee Lee, *Senior Member, IEEE*

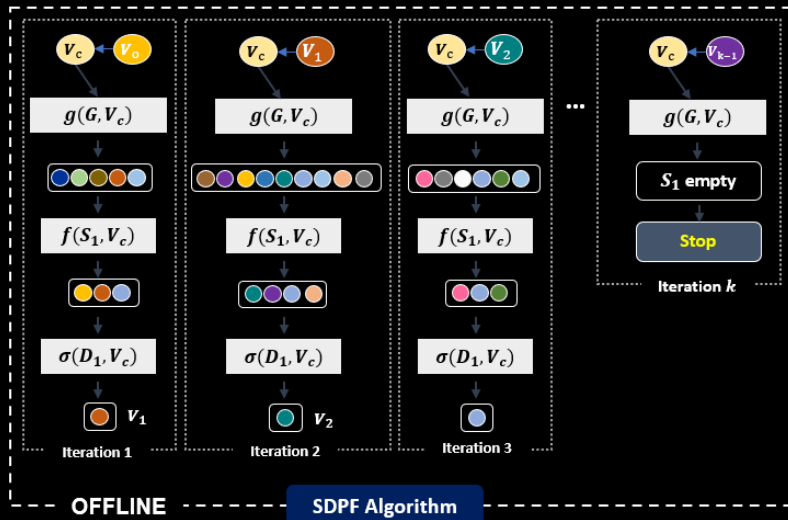
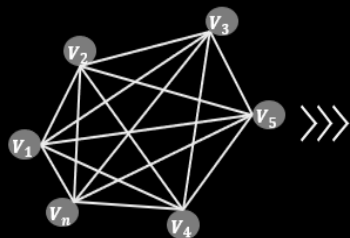
Regenerating arbitrary video sequences with Distillation Path-Finding



Regenerating arbitrary video sequences with Distillation Path-Finding



Regenerating arbitrary video sequences with Distillation Path-Finding



Algorithm 1 SDPF Algorithm

Input: Set of latent vectors $\{v_i\}$, distance metric $\{d_{ij}\}$

- 1: $\mathcal{V} \leftarrow \{v_i\}, \mathcal{E} \leftarrow \{d_{ij}\}$;
- 2: Construct graph $\mathbf{G} = (\mathcal{V}, \mathcal{E})$;
- 3: $V_o \leftarrow$ user's selection;
- 4: Initialize a list \mathcal{P} to subsequently push the selected node to the path;
- 5: Add V_o to \mathcal{P}
- 6: $V_c \leftarrow V_o$; /* V_c is the node at current state*/
- /* Distillation in the first layer*/
- 7: **for** each node $V_j \in \mathbf{G}(\mathcal{V} - \mathcal{P})$ **do**
- 8: **if** $e_{c,j} < \eta$ **then** /* η is defined in Eq.(8)*/
- 9: Add V_j to S_1
- 10: **end if**
- 11: **end for**
- /* Distillation in the second layer */
- 12: **for** each node $V_k \in S_1$ **do**
- 13: **if** $V_c \in LMS$ **then**
- 14: $S_2 = C_d(V_c, V_k) + C_t(V_c, V_k)$
- 15: **else**
- 16: $S_2 = C_t(V_c, V_k)$
- 17: **end if**
- 18: **end for**
- 19: **for** each $V_i \in S_2$ **do**
- 20: Compute possibility Ω for each V_i by Eq.(20);
- 21: **end for**
- 22: Choose V_i by randomly selecting Ω ;
- 23: Add V_i to path \mathcal{P} ;
- 24: Update $V_c \leftarrow V_i$

Output: Sequence of path \mathcal{P}

Regenerating arbitrary video sequences with Distillation Path-Finding

Directional constraint

Coherent constraint

$$\mathcal{S}_2 = \begin{cases} \mathbf{C}_d(V_c, V_k) + \mathbf{C}_t(V_c, V_k), & \text{if } V_c \in LMS \\ \mathbf{C}_t(V_c, V_k), & \text{otherwise} \end{cases}$$

Directional distillation constraint C_d

Case 1: Motions in source video are entirely linear



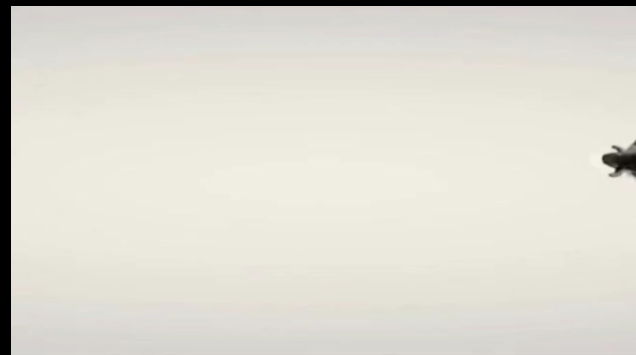
Original video

Rendered clip without constraint C_d

Rendered clip with full configuration

Directional distillation constraint C_d

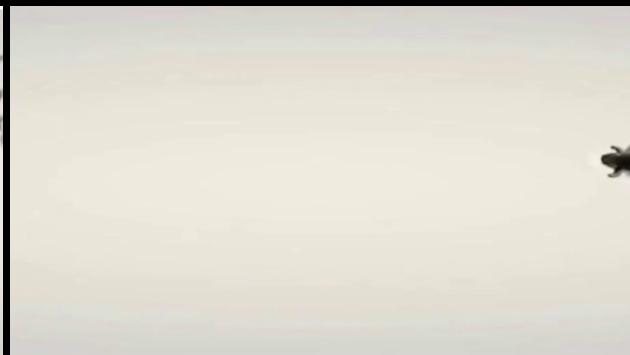
Case 2: Source video with many linear motion segments



Original video



Rendered clip without constraint C_d



Rendered clip with full configuration

▲ Coherent distillation constraint C_t



Original video



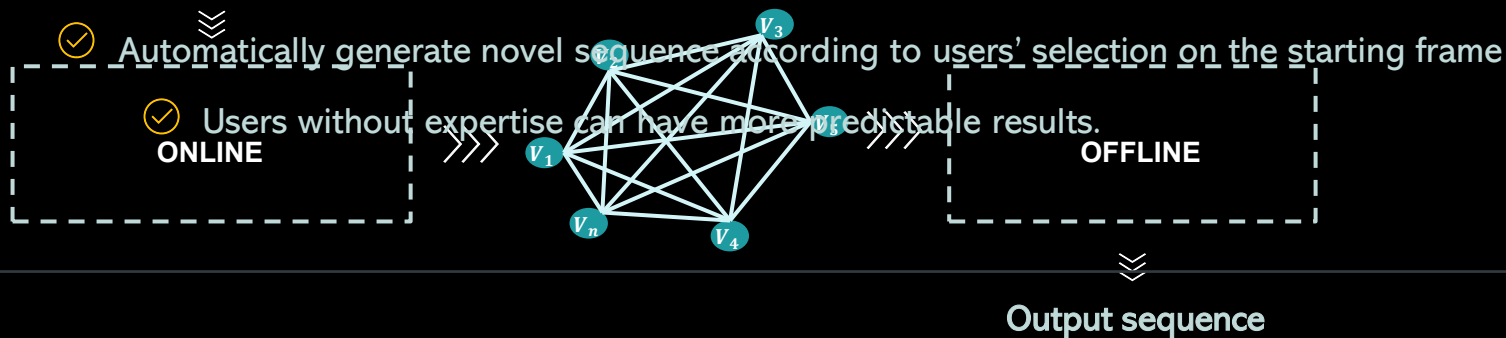
Rendered clip without constraint C_t



Rendered clip with full configuration

SDPF - Benefits

- ✓ Handle both cartoon and natural scenes
- ✓ Work well with dense motions and multiple moving objects



△ Advantage 1: Generating smooth sequence at arbitrary frame



Source video



Source sequence starting at frame 426 (*)



Rendered clip with starting at frame 426

(*) On total 480 frames

△ Advantage 2: Start at the same frame but resulting different sequences

Top right is a sample segment that visualizes the difference in three videos



Original sequence at frame 0



Our system serves different sequences at frame 0



Advantage 3: Control consistency of motion of background and foreground



Original video



Bad result without
our method



Backward motion



Forward motion

Advantage 4: Real-world videos with dense of moving objects



Original video; top-left is bad result without our method



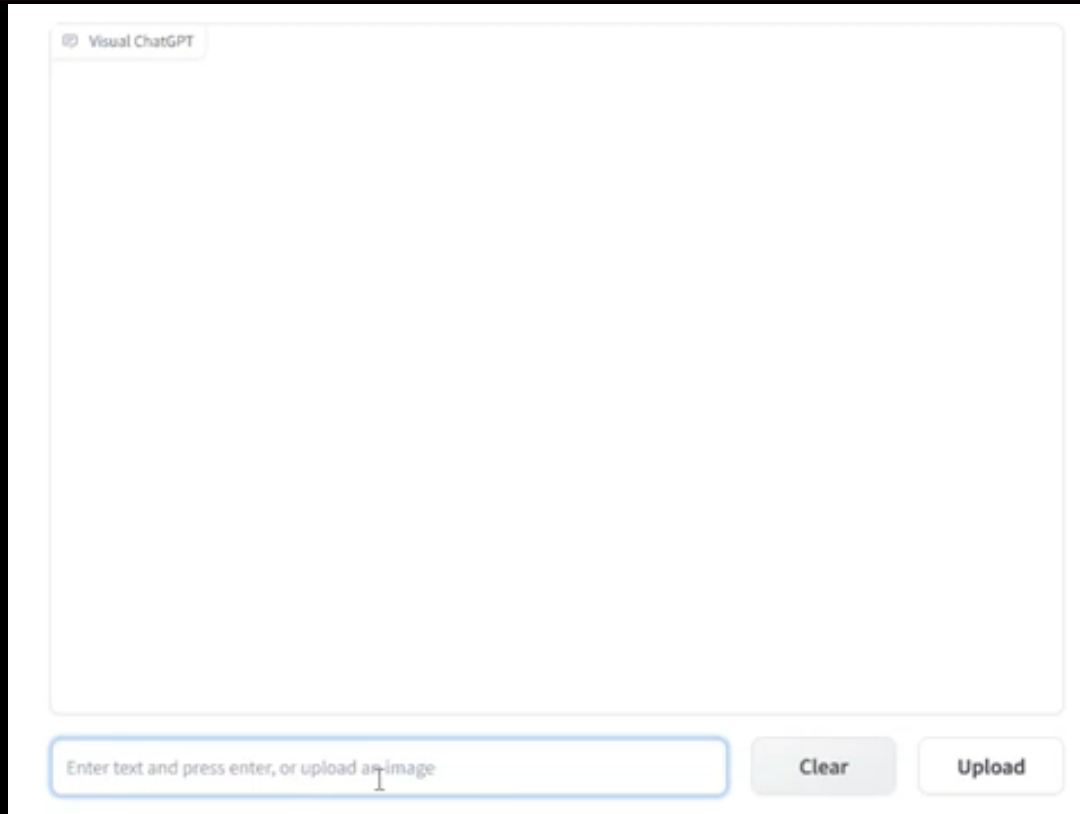
Rendered result by our method

▲ State-of-the-art generation of Animation



Text-driven ????

State-of-the-art generation of Animation



Visual ChatGPT

Text-driven animation

MDM: Human Motion Diffusion Model

ICLR2023 (Top-25%)

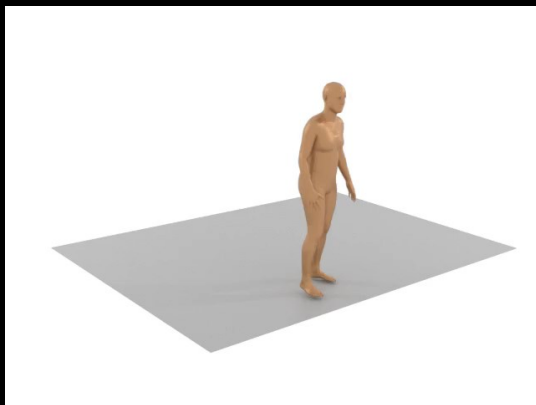
Guy Tevet, Sigal Raab, Brian Gordon, Yonatan Shafir, Daniel Cohen-Or, Amit H. Bermano

Tel Aviv University, Israel

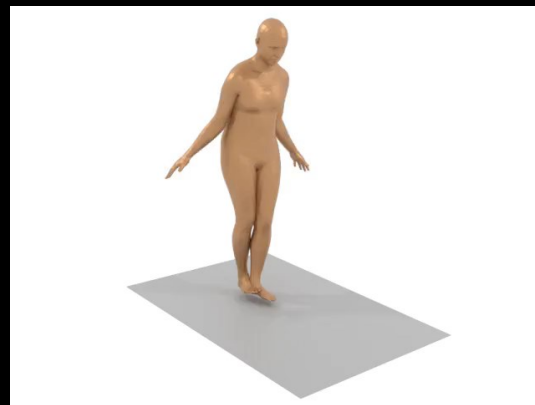
arXiv

Code

Demo



A person turns to his right and paces back and forth



A person is skipping rope

References

1. de Juan, Christina, and Bobby Bodenheimer. "Cartoon textures." In *Proceedings of the 2004 ACM SIGGRAPH/Eurographics symposium on Computer animation*, pp. 267-276. 2004.
2. Yang, Yi, Yueting Zhuang, Dacheng Tao, Dong Xu, Jun Yu, and Jiebo Luo. "Recognizing cartoon image gestures for retrieval and interactive cartoon clip synthesis." *IEEE Transactions on Circuits and Systems for Video Technology* 20, no. 12 (2010): 1745-1756.
3. Yu, Jun, Dongquan Liu, Dacheng Tao, and Hock Soon Seah. "On combining multiple features for cartoon character retrieval and clip synthesis." *IEEE Transactions on Systems, Man, and Cybernetics, Part B (Cybernetics)* 42, no. 5 (2012): 1413-1427.
4. Morace, Charles C., Thi-Ngoc-Hanh Le, Sheng-Yi Yao, Shang-Wei Zhang, and Tong-Yee Lee. "Learning a perceptual manifold with deep features for animation video resequencing." *Multimedia Tools and Applications* 81, no. 17 (2022).
5. Thi-Ngoc-Hanh Le, Sheng-Yi Yao, Chun-Te Wu, and Tong-Yee Lee. "Regenerating Arbitrary Video Sequences with Distillation Path-Finding." *IEEE Transactions on Visualization and Computer Graphics* (2023).
6. Tevet, Guy, Sigal Raab, Brian Gordon, Yonatan Shafir, Daniel Cohen-Or, and Amit H. Bermano. "Human motion diffusion model." *arXiv preprint arXiv:2209.14916* (2022).



▲ End.

Thanks for your listening.